# TSDT14 Signal Theory

Lecture 1
Introduction and Repetition

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# TSDT14 Signal Theory - Formalia

Information & course material: www.commsys.isy.liu.se/TSDT14

Lecturer & examiner: Mikael Olofsson, mikael.olofsson@liu.se

Tutorials: Christopher Mollén, christopher.mollen@liu.se

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Tan Tai Do, tan.tai.do@liu.se

Examination: Laborations (2hp):

Study 1, 2, 3, 4 (4x2 hours)

Sign-up on the web

Report before end of exam period

Written exam (4hp):

1 simple task – Demand: 2/3 OK 5 tasks (5 points each), max 25

Pass: 10 points



# Course Aims 1(2)

After passing the course, the student should

- be able to clearly define central concepts regarding stochastic processes, using own words. *(task 1)*
- be able to reliably perform standard calculations regarding stochastic processes, e.g. LTI filtering (both time continuous and time discrete), sampling and pulse amplitude modulation. (task 1)
- be able to reliably perform standard calculations regarding stochastic processes being exposed to certain momentary non-linearities that are common in telecommunication, especially uniform quantization and monomial non-linearities of low degrees. (task 1)
- with some reliability be able to solve problems that demand integration
  of knowledge from different parts of the course.
  (tasks 2-6)



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# Course Aims 2(2)

After passing the course, the student should

- be able to account for the connection between different concepts in the course in a structured way using adequate terminology.
   (lab report)
- be able to estimate the auto correlation function and power spectral density of a stochastic process based on a realization of the process. Also, clearly and logically account for those estimations and conclusions that can be drawn from them. (*lab report*)



# Languages in Tutorial Sessions (Lessons)

Three tutorial tracks:

- Group A: Christopher Mollén, in Swedish
- Group B: Mikael Olofsson, in Swedish
- Group C: Tan Tai Do, in English

You are free to follow any tutorial group you wish.

You cannot demand language changes for the group teaching.

Individual teaching can be done in any language that works.

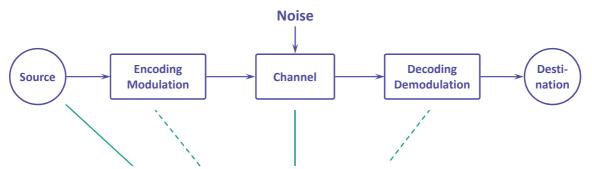


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#### Standard Situation



Partly or completely unknown  $\Rightarrow$  Probabilistic model

We can have Linear and non-linear filtering

Sampling and reconstruction

Up- and down-sampling

Modulation

Also: Error Correction, Packing, Cryptology,...



# Classification of Signals

Deterministic

 $x(t) = e^{-t^2}$ 

Stochastic

Unknown signals noise

Signal Theory

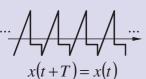
Onedimensional

Multidimensional

x(t) x(a,b,c)

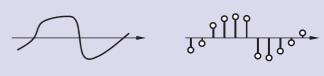
Image Processing

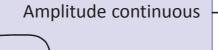
Periodic - Non-periodic





Time-continuous – Time-discrete





- Amplitude discrete





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# Classification of Systems

Linearity – Preserves linear combinations.

Time-invariance – Behaves the same way all the time.

Causality – Has no knowledge about the future.

Stability – Bounded input results in bounded output.

LTI – Linear and Time-Invariant. Convolution.

# Frequency Domain

Time-Continuous Fourier transform:

Signal: x(t)

Spectrum:  $X(f) = F\{x(t)\} = \int_{-\infty}^{\infty} x(t)e^{-j2\pi ft}dt$ Inverse:  $x(t) = F^{-1}\{X(f)\} = \int_{-\infty}^{\infty} X(f)e^{j2\pi ft}df$ 



Jean Baptiste Joseph Fourier 1768 - 1830

Time-Discrete Fourier transform:

Signal: x[n]

Spectrum:  $X[\theta] = F\{x[n]\} = \sum_{n} x[n]e^{-j2\pi\theta n}$ 

Inverse:  $x[n] = F^{-1}\{X[\theta]\} = \int_0^1 X[\theta] e^{j2\pi\theta n} d\theta$ 



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### Signal Power and Signal Energy — Parseval

Signal power:  $|x(t)|^2$ 

Signal energy:  $\int |x(t)|^2 dt$ 

Parseval's relation (special case):

$$\int_{-\infty}^{\infty} |x(t)|^2 dt = \int_{-\infty}^{\infty} x(t) x^*(t) dt = \int_{-\infty}^{\infty} X(f) X^*(f) df = \int_{-\infty}^{\infty} |X(f)|^2 df$$

Energy spectrum:  $|X(f)|^2$ 

 $\int_{-\infty}^{\infty} a(t)b^*(t) dt = \int_{-\infty}^{\infty} A(f)B^*(f) df$ Parseval's relation (generally):

# Output from an LTI System

Notation: 
$$A(f) = \mathcal{F}\{a(t)\}$$
  $B(f) = \mathcal{F}\{b(t)\}$ 

Property: 
$$\mathscr{F}\{(a*b)(t)\} = \int_{-\infty}^{\infty} (a*b)(t) e^{-j2\pi f t} dt = A(f)B(f)$$

LTI System:



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# The Principle sine in – sine out

For LTI systems, we have:

Sine in – Sine out (the same frequency)

More precisely:

Input: 
$$x(t) = \hat{X} \sin(2\pi f_0 t + \varphi)$$

Output: 
$$y(t) = \hat{X} |H(f_0)| \sin(2\pi f_0 t + \varphi + \arg\{H(f_0)\})$$

Amplitude characteristic: |H(f)|

Phase characteristic:  $arg\{H(f)\}$ 

This is the  $j\omega$  method in condensed form.



#### Classification of Frequency Selective Filters

- A (frequency selective) filter is an LTI system. Usually it lets some frequency band through or stops it.
- Usually an electrical network either passive or active.

Notation

Ideal amplitude characteristics

Real amplitude char.

Lowpass filter (LP filter):

Highpass filter (HP filter):

Bandpass filter (BP filter):

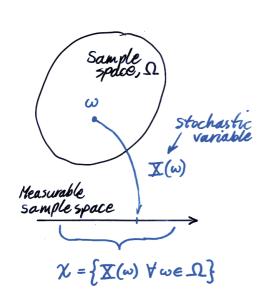
Allpass filter (AP filter):



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#### **Probabilities and Distributions**



Probability:  $Pr\{A\} \in [0,1]$ 

Joint prob.:  $Pr\{A,B\}$ 

Cond. Prob.:  $Pr\{A | B\} = \frac{Pr\{A, B\}}{Pr\{B\}}$ 

Prob. distr.:  $F_X(x) = Pr\{X \le x\} \in [0,1]$ 

Prob. density.:  $f_X(x) = \frac{d}{dx} F_X(x)$ 

**Properties:**  $F_{x}(x)$  is non-decreasing

 $f_X(x) \ge 0$  for all x

 $\int_{-\infty}^{\infty} f_X(x) \ dx = 1$ 

 $\Pr\{x_1 < X \le x_2\} = \int_{x_1^+}^{x_2^+} f_X(x) \ dx$ 

#### Example

Game based on tossing two fair coins:

$$\Omega = \{HH, HT, TH, TT\}$$

$$\Omega_{\mathbf{x}} = \{-200, -100, 400\}$$

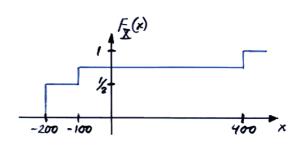
$$X(\omega) = \begin{cases} 400 & \omega = HH \\ -100 & \omega = TT \\ -200 & \omega \in \{HT, TH\} \end{cases}$$

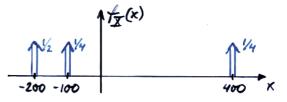
$$F_{X}(x) = \begin{cases} 0, & x < -200 \\ \frac{1}{2}, & -200 \le x < -100 \\ \frac{3}{4}, & -100 \le x < 400 \\ 1, & x \ge 400 \end{cases}$$

$$= \frac{1}{2}u(x+200) + \frac{1}{4}u(x+100) + \frac{1}{4}u(x-400)$$

$$f_{\mathbf{X}}(x) = \frac{d}{dx} F_{\mathbf{Y}}(x)$$









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# **Expectations**

Expectation (mean):

$$E\{X\} = \int_{-\infty}^{\infty} x f_{X}(x) dx$$

For discrete variables also:

$$E\{X\} = \sum_{i} x_{i} \cdot Pr\{X = x_{i}\}$$

For functions of a variable:

$$Y = g(X)$$

$$E\{Y\} = \int_{X} y \cdot f_{Y}(y) dy$$

$$= \int_{X} g(x) \cdot f_{X}(x) dx$$

ALSO:

Quadratic mean (Power):

$$E\{X^2\} = \int_{-\infty}^{\infty} x^2 f_X(x) dx$$

Variance:

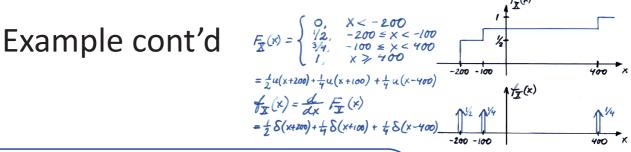
$$Var \{X\} = E\{(X - m_X)^2\}$$

$$= E\{X^2\} - m_X^2$$

ALSO:

Standard deviation:

K



$$Y = sgn X$$
  
 $fr\{Y = 1\} = Pr\{X > 0\} = Pr\{X = 400\} = 1/4$   
 $Pr\{Y = -1\} = Pr\{X < 0\} = Pr\{X = -100\} + Pr\{X = -200\} = 3/4$ 

$$F_{Y}(y) = \frac{3}{4}u(y+1) + \frac{1}{4}u(y-1)$$

$$F_{Y}(y) = \frac{3}{4}\delta(y+1) + \frac{1}{4}\delta(y-1)$$

$$E\{X\} = \int_{-\infty}^{\infty} x \left(\frac{1}{2}\delta(x+200) + \frac{1}{4}\delta(x+100) + \frac{1}{4}\delta(x-400)\right) dx$$

$$= -200 \cdot \frac{1}{2} - 100 \cdot \frac{1}{4} + 400 \cdot \frac{1}{4} = -25$$

$$E\{X\} = (-200)^{2} \cdot \frac{1}{2} + (-100)^{2} \cdot \frac{1}{4} + (400)^{2} \cdot \frac{1}{4} = 62500$$

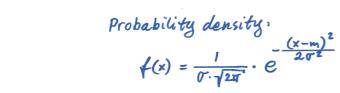
$$Var\{X\} = E\{X^{2}\} - (E\{X\})^{2} = 61875$$



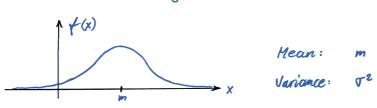
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# Gaussian Distributions, $N(m,\sigma)$



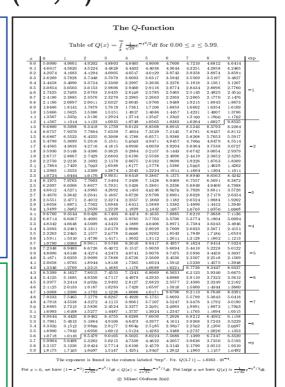
Probability distribution is hard.



The Q function:  

$$Q(x) = 1 - F(x) \text{ for } N(0, 1)$$

$$= \int_{x}^{\infty} \frac{1}{\sqrt{2\pi^{2}}} e^{-t^{2}/2} dt$$
From a table





# Example of the Q Function

$\boldsymbol{x}$	0	1	2	3	4	5	6 ♥	7	8	9	exp
0.0	5.0000	4.9601	4.9202	4.8803	4.8405	4.8006	4.7608	4.7210	4.6812	4.6414	
0.1	4.6017	4.5620	4.5224	4.4828	4.4433	4.4038	4.3644	4.3251	4.2858	4.2465	
0.2	4.2074	4.1683	4.1294	4.0905	4.0517	4.0129	3.9743	3.9358	3.8974	3.8591	
0.3	3.8209	3.7828	3.7448	3.7070	3.6693	3.6317	3.5942	3.5569	3.5197	3.4827	
0.4	3.4458	3.4090	3.3724	3.3360	3.2997	3.2636	3.2276	3.1918	3.1561	3.1207	
0.5	3.0854	3.0503	3.0153	2.9806	2.9460	2.9116	2.8774	2.8434	2.8096	2.7760	-1
0.6	2.7425	2.7093	2.6763	2.6435	2.6109	2.5785	2.5463	2.5143	2.4825	2.4510	-1
0.7	2.4196	2.3885	2.3576	2.3270	2.2965	2.2663	2.2363	2.2065	2.1770	2.1476	
0.8	2.1186	2.0897	2.0611	2.0327	2.0045	1.9766	1.9489	1.9215	1.8943	1.8673	
0.9	1.8406	1.8141	1.7879	1.7619	1.7361	1.7106	1.6853	1.6602	1.6354	1.6109	
1.0	1.5866	1.5625	1.5386	1.5151	1.4917	1.4686	1.4457	1.4231	1.4007	1.3786	
1.1	1.3567	1.3350	1.3136	1.2924	1.2714	1.2507	1.2302	1.2100	1.1900	1.1702	
1.2	1.1507	1.1314	1.1123	1.0935	1.0749	1.0565	1.0383	1.0204	1.0027	9.8525	
1.3	9.6800	9.5098	9.3418	9.1759	9.0123	8.8508	8.6915	8.5343	8.3793	8.2264	
1.4	8.0757	7.9270	7.7804	7.6359	7.4934	7.3529	7.2145	7.0781	6.9437	6.8112	
1.5	6.6807	6.5522	6.4255	6.3008	6.1780	6.0571	5.9380	5.8208	5.7053	5.5917	
1.6	5.4799	5.3699	5.2616	5.1551	5.0503	4.9471	4.8457	4.7460	4.6479	4.5514	_
1.7	4.4565	4.3633	4.2716	4.1815	4.0930	4.0059	3.9204	3.8364	3.7538	3.6727	(-2)
1.8	3.5930	3.5148	3.4380	3.3625	3.2884	3.2157	3.1443	3.0742	3.0054	2.9379	
<b>1.9</b>	2.8717	2.8067	2.7429	2.6803	2.6190	2.5588	2.4998	2.4419	2.3852	2.3295	
2.0	2.2750	2.2216	2.1692	2.1178	2.0675	2.0182	1.9699	1.9226	1.8763	1.8309	
2.1	1.7864	1.7429	1.7003	1.6586	1.6177	1.5778	1.5386	1.5003	1.4629	1.4262	
2.2	1.3903	1.3553	1.3209	1.2874	1.2545	1.2224	1.1911	1.1604	1.1304	1.1011	
2.3	1.0724	1.0444	1.0170	9.9031	9.6419	9.3867	9.1375	8.8940	8.6563	8.4242	

 $Q(1.96) \approx 2.4998 \cdot 10^{-2}$ 



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# Example of the Q Function cont'd

$\boldsymbol{x}$	0	1	2	3	4	5	6	7	8	9	exp
4.2	1.3346	1.2769	1.2215	1.1685	1.1176	1.0689	1.0221	9.7736	9.3447	8.9337	
4.3	8.5399	8.1627	7.8015	7.4555	7.1241	6.8069	6.5031	6.2123	5.9340	5.6675	
4.4	5.4125	5.1685	4.9350	4.7117	4.4979	4.2935	4.0980	3.9110	3.7322	3.5612	(-6)
4.5	3.3977	3.2414	3.0920	2.9492	2.8127	26823	2.5577	2.4386	2.3249	2.2162	
4.6	2.1125	2.0133	1.9187	1.8283	1.7420	1.6597	1.5810	1.5060	1.4344	1.3660	
4.7	1.3008	1.2386	1.1792	1.1226	1.0686	1.0171	9.6796	9.2113	8.7648	8.3391	
4.8	7.9333	7.5465	7.1779	6.8267	6.4920	6.1731	5.8693	5.5799	5.3043	5.0418	
4.9	4.7918	4.5538	4.3272	4.1115	3.9061	3.7107	3.5247	3.3476	3.1792	3.0190	-7
5.0	2.8665	2.7215	2.5836	2.4524	2.3277	2.2091	2.0963	1.9891	1.8872	1.7903	
5.1	1.6983	1.6108	1.5277	1.4487	1.3737	1.3024	1.2347	1.1705	1.1094	1.0515	

$$Q(x) = 10^{-6} \qquad \Rightarrow \qquad x \approx 4.75$$

For 
$$x > 0$$
, we have  $(1 - x^{-2}) \frac{1}{x\sqrt{2\pi}} e^{-x^2/2} dt < Q(x) < \frac{1}{x\sqrt{2\pi}} e^{-x^2/2} dt$ .

For large 
$$x$$
 we have  $Q(x) \approx \frac{1}{x\sqrt{2\pi}}e^{-x^2/2}dt$ .



#### Other Common Distributions

**Uniform distribution:** 

$$f(x) = \begin{cases} \frac{1}{b-a}, & a \leq x < b \\ 0, & \text{elsewhere} \end{cases}$$
 Mean:  $\frac{a+b}{2}$  Variance:  $\frac{(b-a)^2}{12}$ 

**Exponential distribution:** 

$$f(x) = \frac{1}{c} e^{-x/c} \cdot u(x)$$

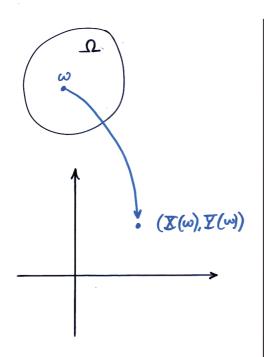
**Binary distribution:** 

$$f(x) = p \cdot \delta(x-a) + (1-p) \cdot \delta(x-b)$$
 Mean:  $p \cdot a + (1-p) \cdot b$   
Vorience:  $p(1-p) \cdot (b-a)^2$ 



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#### Two-Dimensional Stochastic Variables



Distribution: 
$$F_{X,Y}(x,y) = \Pr\{X \leq x, Y \leq y\}$$

Density:  $f_{X,Y}(x,y) = \frac{\partial^2}{\partial x \partial y} F_{X,Y}(x,y)$ 

Properties:  $f_{X,Y}(x,y) \geq 0 \quad \forall x,y$ 

$$\iint_{X,Y} (x,y) \, dx \, dy = 1$$

$$\int_{-\infty}^{\infty} f_{X,Y}(x,y) \, dx \, dy = 1$$
(X(w),Y(w)) Marginal: 
$$f_{X}(x) = \int_{-\infty}^{\infty} f_{X,Y}(x,y) \, dy$$

$$f_{\underline{Y}}(y) = \int_{X} f_{\underline{X},\underline{Y}}(x,y) dx$$

$$f_{\underline{I}}(y) = \int_{\infty}^{\infty} f_{\underline{I},\underline{Y}}(x,y) dx$$
Function:  $E\{g(\underline{I},\underline{I})\} = \int_{\infty}^{\infty} g(x,y) f_{\underline{I},\underline{Y}}(x,y) dx dy$ 

# Dependencies

**Definition:** X & Y are independent if  $F_{X,Y}(x,y) = F_X(x)F_Y(y)$  holds.

 $\Leftrightarrow f_{X,Y}(x,y) = f_X(x) f_Y(y)$  holds. **Theorem:** Independent

**Definition:** Covariance:  $Cov\{X,Y\} = E\{(X-m_X)(Y-m_Y)\}$ 

**Theorem:** Cov $\{X, Y\}$  = E $\{XY\}$  –  $m_x m_y$ 

**Definition:** X & Y are uncorrelated if  $Cov\{X, Y\} = 0$  holds.

**Theorem:** Independent uncorrelated.  $\Rightarrow$ 

Note:  $Var{X} = Cov{X,X}$ 

**Theorem:** Uncorrelated  $\Leftrightarrow$   $E\{XY\} = E\{X\}E\{Y\}$ 

 $\Leftrightarrow$  Var{X+Y} = Var{X} + Var{Y}



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# Bayes' Rule

Conditional distribution: 
$$F_{X|Y}(x|y) = \frac{\int_{\infty}^{x} f_{X,Y}(z,y) dz}{f_{Y}(y)}$$

Conditional density: 
$$f_{\mathbf{x}|\mathbf{y}}(\mathbf{x}|\mathbf{y}) = \frac{d}{d\mathbf{x}} f_{\mathbf{x},\mathbf{y}}(\mathbf{x},\mathbf{y}) = \frac{\forall \mathbf{x},\mathbf{y}(\mathbf{x},\mathbf{y})}{\forall \mathbf{y}(\mathbf{y})}$$

Bayes rule (continuous): 
$$f_{X|X}(x|y) = \frac{f_{X|X}(y|x)}{f_{X}(y)} \cdot f_{X}(x)$$

(discrete): 
$$Pr\{X=x|Y=y\} = \frac{Pr\{Y=y|X=x\}}{Pr\{Y=y\}} \cdot Pr\{X=x\}$$
  
(Mixed):  $Pr\{X=x|Y=y\} = \frac{f_{X}[X](y|x)}{f_{X}(y)} \cdot Pr\{X=x\}$ 

(Mixed): 
$$Pr\{X=x|Y=y\} = \frac{f_{X|X}(y|x)}{f_{Y}(y)} \cdot Pr\{X=x\}$$

#### Multi-Dimensional Stochastic Variables

Distribution: 
$$F_{X_1,...,X_N}(x_1,...,x_N) = Pr\{X_1 \leq x_1,...,X_N \leq x_N\}$$

Density: 
$$f_{X_1,...,X_N}(x_1,...,x_N) = \frac{\partial^N}{\partial x_1...\partial x_N} f_{X_1,...,X_N}(x_1,...,x_N)$$

Vector notation: 
$$\overline{\mathbf{X}} = (\mathbf{X}_1, ..., \mathbf{X}_N), \overline{\mathbf{X}} = (\mathbf{X$$

You might think that this makes sense:

Independence: 
$$f_{\overline{x}}(\overline{x}) = \prod_{i=1}^{N} f_{\overline{x}_{i}}(x_{i})$$
 &  $f_{\overline{x}}(\overline{x}) = \prod_{i=1}^{N} f_{\overline{x}_{i}}(x_{i})$ 

Uncorrelated: 
$$E\{\prod_{i=1}^{N} X_i\} = \prod_{i=1}^{N} E\{X_i\}$$
  
 $Var\{\sum_{i=1}^{N} X_i\} = \sum_{i=1}^{N} Var\{X_i\}$ 

Not entirely true. See course book for details (p. 58-59)



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# Jointly Gaussian Variables

\(\overline{\textbf{X}}=(\overline{\textbf{X}},...,\overline{\textbf{X}}\_N)\) is called jointly Gaussian if the following holds:

$$\oint_{\overline{\mathbf{x}}} (\overline{\mathbf{x}}) = \frac{1}{(2\pi)^{N/2} |\Lambda|^{1/2}} \cdot e^{-\frac{1}{2}(\overline{\mathbf{x}} - \overline{\mathbf{m}}) \Lambda^{-1} (\overline{\mathbf{x}} - \overline{\mathbf{m}})^{T}}$$

$$\overline{m} = E\{\overline{X}\} \qquad \Lambda = \begin{pmatrix} \lambda_{11} \cdots \lambda_{1N} \\ \vdots & \vdots \\ \lambda_{N1} \cdots \lambda_{NN} \end{pmatrix} \qquad \lambda_{ij} = Cov\{X_i, X_i\}$$

$$\lambda_{ij} = Cov\{X_i, X_i\}$$

If 
$$X_1,...,X_N$$
 are pairwise uncorrelated.  $\Rightarrow \Lambda = \begin{pmatrix} O_{X_1}^2 & O \\ O & O_{X_N}^2 \end{pmatrix} \Rightarrow$ 

$$\Rightarrow \sqrt{\bar{\mathbf{x}}}(\bar{\mathbf{x}}) = \frac{1}{(2\pi)^{N/2} \cdot \tilde{\mathbf{y}} \cdot \tilde{\mathbf{c}}_{\bar{\mathbf{x}}_i}} \cdot e^{-\frac{1}{2} \sum_{i} \frac{(\mathbf{x}_i - \mathbf{m}_{\bar{\mathbf{x}}_i})^2}{\sigma_{\bar{\mathbf{x}}_i}^2}} = \tilde{\mathbf{y}} \sqrt{\mathbf{x}_i} (\mathbf{x}_i)$$

: Independent

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